

1N-66  
176655  
P-7  
5012418

# Minimize System Cost by Choosing Optimal Subsystem Reliability and Redundancy

Ronald C. Suich  
*California State University  
Fullerton, California*

and

Richard L. Patterson  
*Lewis Research Center  
Cleveland, Ohio*

Prepared for the  
Reliability and Maintainability Symposium  
Atlanta, Georgia, January 25-28, 1993

(NASA-TM-106251) MINIMIZE SYSTEM  
COST BY CHOOSING OPTIMAL SUBSYSTEM  
RELIABILITY AND REDUNDANCY (NASA)  
7 p

N93-31145

Unclass



G3/66 0176655

# Minimize System Cost by Choosing Optimal Subsystem Reliability & Redundancy

Ron C. Suich  
California State University  
Fullerton, California 92631

and Richard L. Patterson  
National Aeronautics and Space Administration  
Lewis Research Center  
Cleveland, Ohio 44135

## SUMMARY & CONCLUSIONS

The basic question which we address in this paper is how to choose among competing subsystems. This paper utilizes both reliabilities and costs to find the subsystems with the lowest overall expected cost. The paper begins by reviewing some of the concepts of expected value. We then address the problem of choosing among several competing subsystems. These concepts are then applied to k-out-of-n:G subsystems. We illustrate the use of the authors' basic program in viewing a range of possible solutions for several different examples. We then discuss the implications of various solutions in these examples.

## 1. INTRODUCTION

How does a design engineer or manager choose between a power subsystem with .990 reliability and a more costly subsystem with .995 reliability? When is the increased cost of a more reliable subsystem justified?

High reliability is not necessarily an end in itself. High reliability may be desirable in order to reduce the statistically expected cost due to a subsystem failure. However, this may not be the wisest use of funds since the expected cost due to subsystem failure is not the only cost involved. To answer this question the engineer needs to consider not only the cost of the subsystem but also to assess the costs that would occur if the subsystem fails. These costs are weighted by the probability of their occurrence to yield the expected cost. We therefore minimize the total of the two costs, i.e., the total of the cost of the subsystem plus the expected cost due to subsystem failure.

Since this problem involves probabilistic decision making, we'll first review some aspects of probability and expected value. We'll then apply these procedures to a simple situation of choosing between two or more competing subsystems and show how to choose the best subsystem. These principles will then be applied to choosing from among various k-out-of-n:G subsystems. The authors have written a basic program (CARRAC) which enables the engineer to explore and graph various options. We'll illustrate the use of this program with several different cost models.

## Notation

- n number of modules in the subsystem
- k minimum number of good modules for the subsystem to be good
- r reliability of the whole system for other than failure of the subsystem
- $c_1$  loss due to failure of the subsystem
- $c_3$  cost of a one module subsystem capable of full output
- $c_4$  cost of a module in a k-out-of-n:G subsystem when k is fixed
- $r_{si}$  reliability of subsystem i,  $i = 1, 2, \dots$
- $g(k)$  function relating cost of subsystem to the number of modules in subsystem
- p probability that a module is good
- q probability that a module fails or 1-p
- C the total of the cost of the subsystem itself plus the expected cost due to subsystem failure

## 2. EXPECTED VALUE

Since much of the paper is founded upon the idea of expected value or expected cost, we'll review some of the fundamental uses of this concept in decision-making applications.

Suppose that you may choose between actions A and B. In this example, action A always results in a \$1000 return to you. Then A has a value of \$1000 and we can say that the expected value of A,  $E(A)$ , is \$1000. Action

B, on the other hand, results in a return to you of either \$500, outcome  $B_1$ , or \$1500, outcome  $B_2$ . This return is a random variable which depends upon circumstances beyond your control. The choices which you face are outlined in figure 1.

If  $B_1$  and  $B_2$  are equally likely, i.e.,  $\Pr(B_1) = \Pr(B_2) = .5$

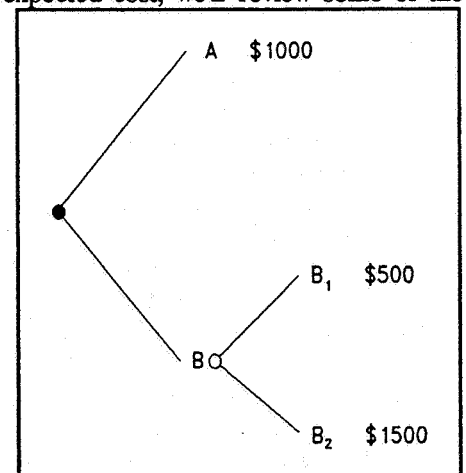


Figure 1 Example of Expected Value.

(where Pr means "probability of"), then  $E(B) = \$500 \times \text{Pr}(B_1) + \$1500 \times \text{Pr}(B_2) = \$500(.5) + \$1500(.5) = \$1000$ . If you use expected value as your criterion, then you would be indifferent as to choice A or B, since both have an expected value of \$1000. Also note that, although B has an expected value of \$1000, you never receive \$1000. Half of the time you receive \$500 and half of the time you receive \$1500.

Now suppose that the probabilities of  $B_1$  and  $B_2$  are .4 and .6, respectively. Then  $E(B) = \$500(.4) + \$1500(.6) = \$1100$ . If you use expected value as your criterion, you would choose B over A, since it has the higher expected value. In unusual circumstances, such as the need to repay \$1000, you might choose A over B, even though A has the lower expected value. For these types of circumstances, we say that the certain return of \$1000 has a higher expected utility to you than the expected utility associated with an expected value of \$1100, where the return can be either \$500 or \$1500. For unusual circumstances, the procedures outlined in this paper can be applied using expected utility rather than expected value. For a more detailed discussion of utility, see [1].

Suppose instead that action A results in a cost of \$1000 while action B results in a cost of either \$500 or \$1500. We could, in a manner similar to that above, analyze actions A and B in terms of their expected costs. Our objective would be to minimize expected cost. Throughout the remainder of this paper we will use expected value or expected cost as our criterion.

### 3. BALANCING TWO COSTS

We will be using expected value as our criterion, namely the expected cost due to subsystem failure, shown as  $E\{\text{cost due to subsystem failure}\}$ . As with all expected values, this number depends upon both the dollar cost and the probability of its occurrence. Let  $c_1$  be the dollar cost due to failure of the subsystem, including all costs incurred by subsystem failure (but not the cost of the subsystem itself). This number could be the entire cost of the main system (even greater in some circumstances) if failure of the subsystem resulted in complete failure of the main system. In other instances  $c_1$  could be less than the cost of the subsystem, e.g., cost of the subsystem resulted in only a partial failure of the main system.

Now the expected cost due to subsystem failure is  $c_1$  times the probability that this cost will be experienced. If the main system fails (for other than failure of the subsystem) then there is no cost due to subsystem failure. For example, if we're considering a power subsystem in a rocket, the rocket may explode on the launch pad due to a fuel problem. Even if the power subsystem failed in flight, we would not experience this failure. Let  $r$  be the reliability of the main system (for other than failure of the subsystem) and let  $r_s$  be the reliability of the subsystem. Then  $E\{\text{cost due to subsystem failure}\} = c_1 \text{Pr}\{\text{subsystem failure} \mid \text{main system good}\} \text{Pr}\{\text{main system good}\} = c_1(1-r_s)r = rc_1(1-r_s)$ .

We can minimize this expected cost by building a

subsystem with an extremely low probability of failure, i.e., a subsystem with extremely high reliability. In this situation it is not clear that we should build the most reliable subsystem possible since this will minimize only the expected cost due to subsystem failure but does not consider the cost of the building the subsystem itself. To make this decision, we should not consider the two costs separately. We therefore minimize the total of the two costs, i.e., the total of the cost of the subsystem plus the expected cost due to subsystem failure. The total quantity to be minimized is given by

$$\begin{aligned} C &= \text{cost of the subsystem} + E\{\text{cost due to subsystem failure}\} \\ &= \text{cost of the subsystem} + rc_1(1-r_s). \end{aligned}$$

In minimizing C, we see that we are balancing the cost of the subsystem itself and the expected cost due to subsystem failure.

### 4. SELECTING THE BETTER SUBSYSTEM

As an example, suppose that we have two possible subsystems under consideration. Subsystem 1, which costs \$200,000, has a .97 reliability. Subsystem 2, with a cost of \$100,000, has a .94 reliability. Without further information and analysis, there is no clear "best" subsystem, and the choice is often based upon the amount budgeted for the subsystem.

For further analysis, let's assume that the two subsystems under consideration will be part of a main system which has a reliability (exclusive of the subsystem under consideration) of  $r = .96$ . We'll further assume that failure of the subsystem will result in a cost of  $c_1 = \$10,000,000$ . Let us first look at the  $E\{\text{cost due to subsystem failure}\}$  for each of the two subsystems. For subsystem 1,

$$\begin{aligned} E\{\text{cost due to subsystem failure}\} &= rc_1 \text{Pr}\{\text{subsystem failure}\} \\ &= rc_1(1-r_{s1}) = .96 \times \$10,000,000 \times .03 = \$288,000. \end{aligned}$$

For subsystem 2,

$$\begin{aligned} E\{\text{cost due to subsystem failure}\} &= rc_1(1-r_{s2}) = \\ &= .96 \times \$10,000,000 \times .06 = \$576,000. \end{aligned}$$

Since subsystem 2 is less reliable than subsystem 1 it has a higher expected cost. However, since 2 is also less expensive, we need to compare the overall expected cost,  $C$ , for 1 and for 2. For subsystem 1,

$$C_{s1} = \$200,000 + \$288,000 = \$488,000.$$

For subsystem 2,

$$C_{s2} = \$100,000 + \$576,000 = \$676,000.$$

Since  $C_{s1} < C_{s2}$ , we select subsystem 1 over subsystem 2.

### 5. K-OUT-OF-N:G SUBSYSTEMS

We'll now direct our attention to a specific type of subsystem, called a k-out-of-n:G subsystem. Such a subsystem has  $n$  modules, of which  $k$  are required to be good for the subsystem to be good. As an example consider the situation where the engineer has a certain power requirement. He may meet this requirement by having one large power module, two smaller modules, etc. The number of modules required is called  $k$ . For example, the engineer

may decide that  $k = 4$ . Then each module is  $1/4$  of the full required power. Therefore, the subsystem must have 4 or more modules for the full required power. The number of modules used in the subsystem is called  $n$ . For example, an  $n = 6$  and  $k = 4$  subsystem would have 6 modules each of  $1/4$  power and thus would have the output capability of 1.5 times the required power. The engineer chooses  $n$  and  $k$ . Selection of the different values of  $n$  and  $k$  results in different subsystems, each with different costs and reliabilities. Since each  $n$  and  $k$  yields different subsystems with different costs, we can choose the subsystem (the  $n$  and  $k$ ) which will minimize cost  $C$ .

#### Assumptions for k-out-of-n:G subsystems

1. The probability of failure of any module in the system is not affected by the failure of any other module, i.e., the modules are s-independent.
2. There is a k-out-of-n:G subsystem where each of the modules has the same probability of success.
3. Failure of the subsystem results in a loss of  $c_1$ ;  $c_1$  includes all losses incurred due to subsystem failure but not the cost of the subsystem itself.

#### 5.1 MODEL 1

For model 1 we assume that  $k$  is fixed and that each module costs  $c_4$ . Now  $E\{\text{cost due to subsystem failure}\} = rc_1 \Pr\{\text{subsystem failure}\} = rc_1 \Pr\{X < k\} = rc_1 \text{binf}(k-1; p, n)$ . Since  $C = \text{cost of subsystem} + E\{\text{cost due to subsystem failure}\}$ , then  $C = nc_4 + rc_1 \text{binf}(k-1; p, n)$ .

The authors have written a Basic program (CARRAC) to find the  $n$  which minimizes  $C$ . Additionally CARRAC will graph  $C$  as a function of either  $p$  or  $c_1$  and graphs the best subsystems (i.e. the ones with the lowest  $C$ 's) over ranges of  $p$  or  $c_1$ . This allows you to not only select the best subsystem for a particular value of  $p$  or  $c_1$  but also to view what happens to  $C$  for nearby values of  $p$  or  $c_1$ .

As an example, consider the situation when  $k = 1$ , that is only one module is required to be good for the subsystem to be good. The reliability of this single module is estimated to be .95 ( $p = .95$ ). Let the reliability of the system for other than failure of the subsystem be .9 ( $r = .9$ ). The cost of one module is 1 ( $c_4 = 1$ ) million dollars. The cost due to failure of this subsystem is 10 ( $c_1 = 10$ ) million dollars.

Figure 2 shows a plot of  $C$  for  $.79 < p < .99$  and  $n$ 's of 1 through 4. When the reliability of a single module is  $p = .95$ , then the  $n = 1$  subsystem has the lowest value of  $C$ . Therefore the best subsystem is the one with no spares. We see from figure 2 that the  $n = 1$  subsystem has the lowest value of  $C$  for any  $p > .87$ . If  $p < .87$ , then  $n = 2$  (one spare) has the lowest value of  $C$ .

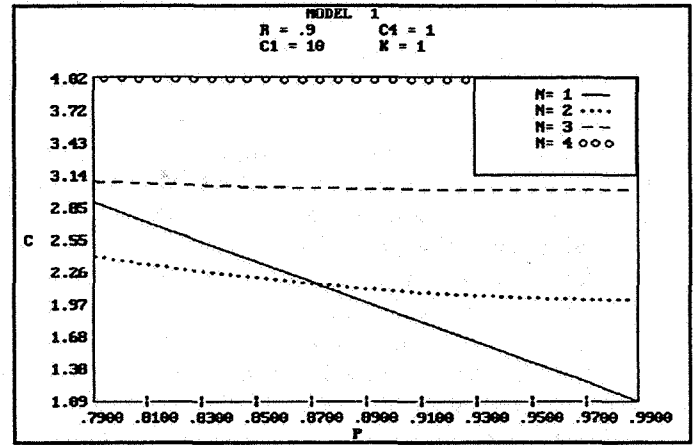


Figure 2 {Cost of Subsystem plus Expected Cost Due to Subsystem Failure} vs Reliability of a Single Subsystem Module.

Now suppose instead that  $c_1$  (cost due to failure of the subsystem) is 50. Figure 3 shows the plot of  $C$  for  $c_1 = 50$ . We first note that if  $p = .95$ , then the  $n = 2$  subsystem (one spare) is the best. Comparing figures 2 and 3 (at  $p = .95$ ) we see that the larger value of  $c_1$  (in figure 3) requires a larger value of  $n$ . In general, if the cost of subsystem failure increases, then more redundancy is required. If  $.83 < p < .98$ , figure 3 shows that the  $n = 2$  subsystem is best. If  $p < .83$  then still more redundancy ( $n=3$ ) is required. If  $p > .98$ , then no redundancy ( $n=1$ ) is required.

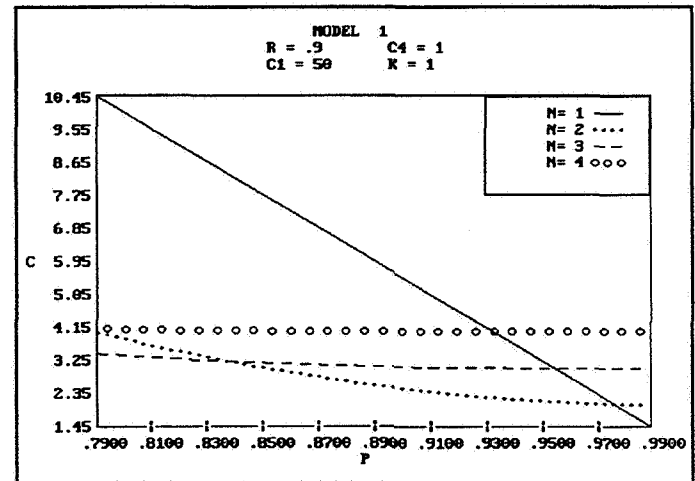


Figure 3 {Cost of Subsystem plus Expected Cost Due to Subsystem Failure} vs Reliability of a Single Subsystem Module.

#### 5.2 MODEL 2

##### Assumptions

Same as model 1 except:

1. We are free to choose  $k$  in our subsystem.
2. The cost of a  $k$ -module subsystem is  $c_3 g(k)$ .
3. Each module in the subsystem costs  $c_3 g(k)/k$ . Since there are  $n$  modules in the subsystem, the cost of the subsystem =  $nc_3 g(k)/k$ . Therefore  $C$



$$= \text{cost of subsystem} + E\{\text{loss due to subsystem failure}\} = nc_3g(k)/k + rc_1 \text{binf}(k-1; p, n).$$

We note that  $g(k)$  usually increases in  $k$ , since it is generally more expensive to have subsystems consisting of  $k$  smaller elements than to have a subsystem consisting of a single large module. As an example of model 2, suppose we are building a space electrical power subsystem. The cost due to subsystem failure,  $c_1$ , is 240. Let the reliability of the system for other than failure of the subsystem be .9 ( $r = .9$ ). Suppose that the cost of building a single module capable of full power is 1 ( $c_3 = 1$ ). A rough rule of thumb says that the cost of smaller modules for a space electrical power subsystem is proportional to the electrical power raised to the .7, i.e.,  $g(k) = k(1/k)^{.7}$ . Therefore, a subsystem consisting of a single module capable of full power would cost  $c_3g(1) = c_31(1/1)^{.7} = 1.0c_3$ , a subsystem consisting of 2 modules, each of 1/2 power, would cost  $c_3g(2) = c_32(1/2)^{.7} = 1.23c_3$  to build, etc. An  $n = 3$  and  $k = 2$

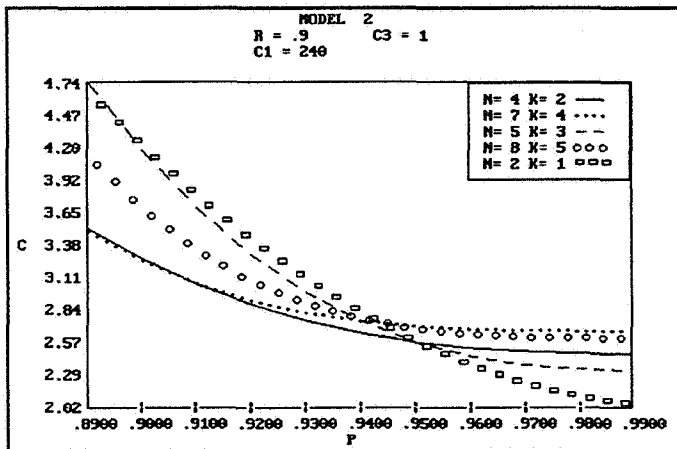


Figure 4 {Cost of Subsystem plus Expected Cost Due to Subsystem Failure} vs Reliability of a Single Subsystem Module.

subsystem, i.e., one having 3 modules each of 1/2 power, would cost  $nc_3g(k)/k = 3 \times 1.23c_3/2 = 1.85c_3$  to build. An estimate of  $p$ , the reliability of an individual module, is .96. If we are unsure of this estimate, we can use CARRAC to view (figure 4) the best subsystems over  $p$  ranging from .89 to .99. From figure 4, at  $p = .96$ , the  $n = 2, k = 1$  subsystem is best (lowest value of  $C$ ). If  $p < .95$ , the  $n = 4, k = 2$  subsystem is best. Note this is a flatter curve over the range of  $p$ , indicating a low value for  $C$  over a wide range of  $p$ .

For the example, suppose we wish to view what happens to  $C$  as  $c_1$  varies. Possibly we are fairly confident about our estimate of  $p = .96$  but unsure about our estimate of  $c_1$ . Figure 5 shows, if  $c_1$  is below 310, that the  $n = 2, k = 1$  subsystem is best. However, for  $310 < c_1 < 400$ , the  $n = 5, k = 3$  subsystem is the best. For  $c_1 > 400$  the  $n = 4, k = 2$  subsystem is the best. This type of analysis can be used whenever you are unsure of  $c_1$  and wish to consider a wider range of values.

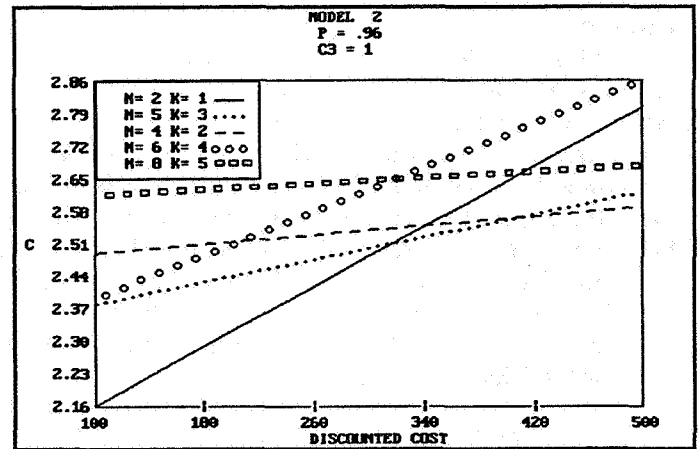


Figure 5 {Cost of Subsystem plus Expected Cost Due to Subsystem Failure} vs (Discounted) Cost of Subsystem Failure.

## 6. OTHER MODELS AND THE BASIC PROGRAM

CARRAC can be used to explore near optimal solutions for the three other cost models presented by Suich & Patterson [2,3]. These other models cover time dependency and situations with and without salvage value. The authors have sent copies of the CARRAC to selected organizations in the United States for initial testing. If you or your organization are interested in participating, please contact Richard Patterson. It is anticipated that CARRAC will be available in the future through NASA's Computer Software Management and Information Center (COSMIC).

## 7. REFERENCES

- [1] Anderson, Sweeney and Williams, Introduction to Management Science, Sixth Edition, West Publishing Company.
- [2] Suich, R. and Patterson, R., "Balancing Reliability and Cost to Choose the Best Power Subsystem", NASA Technical Memorandum 104453, 1991.
- [3] Suich, R. and Patterson, R., "K-out-of-n: G Systems: Some Cost Considerations", IEEE Transactions on Reliability, Vol. 40, No. 3, 1991.
- [4] Suich, R. and Patterson, R., "How Much Redundancy? - Some Cost Considerations, Including Examples for Spacecraft Systems", NASA Technical Memorandum 103197, 1990.

## 8. BIOGRAPHIES

Ron C. Suich  
Management Science Department  
California State University  
Fullerton, CA 92634 USA

Ron C. Suich is Professor of Management Science at California State University, Fullerton. He holds a BBA in marketing from John Carroll

University, and an MSc & PhD in Statistics from Case Western Reserve University. He is (co)author of papers in IEEE Trans. Reliability, J. Royal Statistical Society (Series B), Technometrics, J. Statistical Computation and Simulation, J. Quality Technology, Teaching Statistics, Computational Statistics and Data Analysis, and British J. Mathematical and Statistical Psychology.

Richard L. Patterson  
MS301-2  
NASA Lewis Research Center  
Cleveland, Ohio 44135 USA

Richard L. Patterson is the Program Manager for Fault Tolerant Power Technology and System Diagnostics at the NASA Lewis Research Center. He holds a BS in Physics from John Carroll University, has done graduate work in nuclear/mechanical engineering at Carnegie-Mellon University, and holds an MSc in Engineering Administration from the University of Syracuse. He has 3 patents in diagnostic sensors and systems. His present work includes development of fiber-optic sensors, neural network diagnostic systems, and fuzzy-logic built-in-test. He is a Senior Member of the IEEE.

REPORT DOCUMENTATION PAGE			Form Approved OMB No. 0704-0188	
Public reporting burden for this collection of information is estimated to average 1 hour per response, including the time for reviewing instructions, searching existing data sources, gathering and maintaining the data needed, and completing and reviewing the collection of information. Send comments regarding this burden estimate or any other aspect of this collection of information, including suggestions for reducing this burden, to Washington Headquarters Services, Directorate for Information Operations and Reports, 1215 Jefferson Davis Highway, Suite 1204, Arlington, VA 22202-4302, and to the Office of Management and Budget, Paperwork Reduction Project (0704-0188), Washington, DC 20503.				
1. AGENCY USE ONLY (Leave blank)		2. REPORT DATE January 1993		3. REPORT TYPE AND DATES COVERED Technical Memorandum
4. TITLE AND SUBTITLE  Minimize System Cost by Choosing Optimal Subsystem Reliability and Redundancy			5. FUNDING NUMBERS  WU-506-41-41	
6. AUTHOR(S)  Ronald C. Suich and Richard L. Patterson				
7. PERFORMING ORGANIZATION NAME(S) AND ADDRESS(ES)  National Aeronautics and Space Administration Lewis Research Center Cleveland, Ohio 44135-3191			8. PERFORMING ORGANIZATION REPORT NUMBER  E-7973	
9. SPONSORING/MONITORING AGENCY NAME(S) AND ADDRESS(ES)  National Aeronautics and Space Administration Washington, D.C. 20546-0001			10. SPONSORING/MONITORING AGENCY REPORT NUMBER  NASA TM-106251	
11. SUPPLEMENTARY NOTES Prepared for the Reliability and Maintainability Symposium, Atlanta, Georgia, January 25-28, 1993. Ronald C. Suich, California State University, Fullerton, California 92631; and Richard L. Patterson, NASA Lewis Research Center. Responsible person, Richard L. Patterson, (216) 433-8166.				
12a. DISTRIBUTION/AVAILABILITY STATEMENT  Unclassified - Unlimited Subject Category 66			12b. DISTRIBUTION CODE	
13. ABSTRACT (Maximum 200 words)  The basic question which we address in this paper is how to choose among competing subsystems. This paper utilizes both reliabilities and costs to find the subsystems with the lowest overall expected cost. The paper begins by reviewing some of the concepts of expected value. We then address the problem of choosing among several competing subsystems. These concepts are then applied to k-out-of-n: G subsystems. We illustrate the use of the authors' basic program in viewing a range of possible solutions for several different examples. We then discuss the implications of various solutions in these examples.				
14. SUBJECT TERMS  Cost analysis; Cost; Reliability; Redundancy; Systems Engineering			15. NUMBER OF PAGES 7	
			16. PRICE CODE A02	
17. SECURITY CLASSIFICATION OF REPORT Unclassified	18. SECURITY CLASSIFICATION OF THIS PAGE Unclassified	19. SECURITY CLASSIFICATION OF ABSTRACT Unclassified	20. LIMITATION OF ABSTRACT	